



# In-Vehicle Coupon Recommendation System

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**Github** : <https://github.com/jayshivankar/In-Vehicle-Coupon-Recommendation>

# Project Overview & Methodology

## Problem Statement:

- To develop a highly accurate predictive model that determines whether a driver will accept a coupon.
- The goal is to maximize marketing campaign efficiency by targeting users who are most likely to accept offers.

## Dataset & Preprocessing:

- Utilized a dataset with user, context, and coupon attributes.
- Preprocessing Steps: Managed missing values, performed Label Encoding for categorical features, and used SMOTE to balance the dataset, preventing model bias.

## Methodology: Model Selection

- Comprehensive Evaluation: Trained and evaluated a wide range of classification models:
- Decision Tree, Random Forest, XGBoost, AdaBoost, CatBoost, Logistic Regression, SVM, and KNN.
- Hyperparameter Tuning: Used GridSearchCV to find the optimal parameters for each model, ensuring a fair and robust comparison.
- Champion Model: After a thorough comparison, XGBoost was selected as the final model due to its superior performance across key metrics.







# Exploratory Data Analysis

## (Key Insights)

Overall Coupon Acceptance:

- The dataset shows a relatively balanced distribution, with approximately 57% of coupons being accepted.
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## What Drives Coupon Acceptance?

01.

### Passenger Type:

Drivers with friends as passengers were significantly more likely to accept coupons compared to those driving alone or with family.

02.

### Coupon Type:

Coupons for "Carry out & Take away" and "Restaurants (<\$20)" had the highest acceptance rates. Bar and coffee house coupons were also popular.

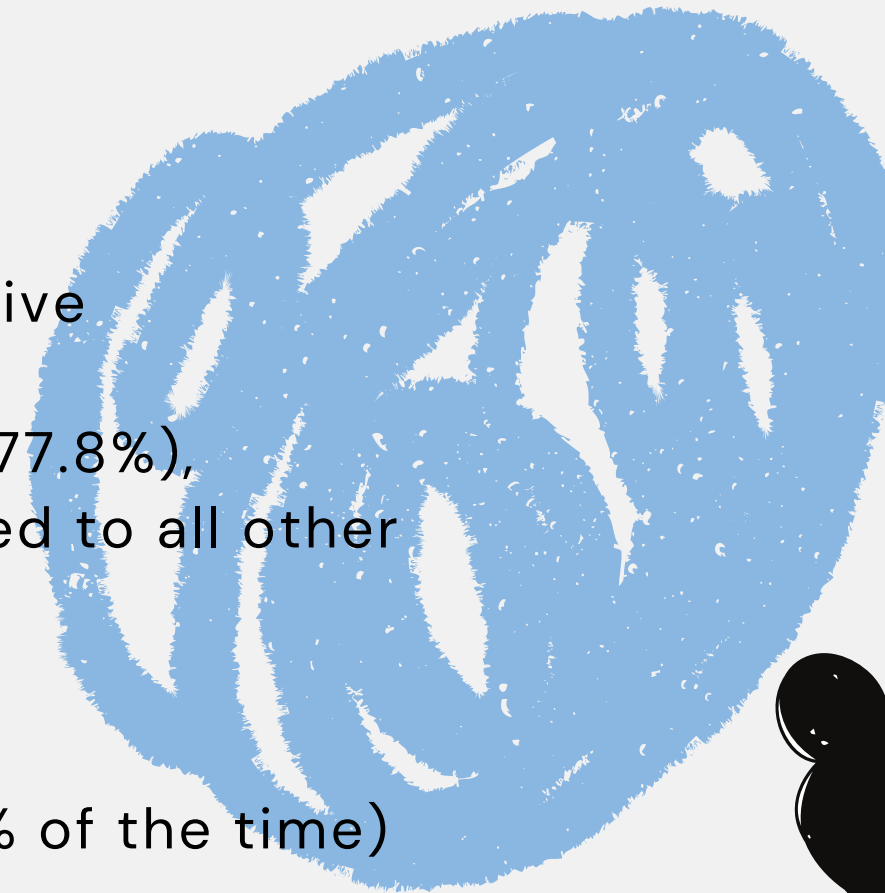
03.

### Occupation:

Mostly the customers who bought the most coupons were either unemployed who had just completed graduation or were students in college or high schools.

# Model Performance: Why XGBoost?

- XGBoost delivered the best overall performance after extensive hyperparameter tuning.
- It achieved the highest Test Accuracy (74.1%) and F1-Score (77.8%), indicating a superior balance of precision and recall compared to all other models, including Random Forest and CatBoost.
- XGBoost Final Performance:
  - Accuracy: 0.7410
  - Precision: 0.7433 (Correctly identifies coupon acceptors 74% of the time)
  - Recall: 0.8169 (Finds 82% of all actual coupon acceptors)
  - F1-Score: 0.7784
- Key Drivers of Prediction (Feature Importance):
- The XGBoost model identified the following as the most influential factors:
  - a.expiration: How soon the coupon expires.
  - b.coupon: The specific type of coupon offered.
  - c.toCoupon\_GEQ25min: Whether the travel time to the location is over 25 minutes.
  - d.Bar: If the coupon is for a bar.



The background is a light gray color, decorated with various hand-drawn blue doodles. These include several loops and swirls at the top, a series of vertical lines on the right side, a wavy line at the bottom center, and several checkmarks at the bottom right. The central text is in a bold, black, sans-serif font with a white drop shadow.

**Thank you  
very much!**